

# General Matrix-Valued Inhomogeneous Linear Stochastic Differential Equations and Applications

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**Abstract:** The expressions of solutions for general  $n \times m$  matrix-valued inhomogeneous linear stochastic differential equations are derived. This generalizes a result of Jaschke (2003) for scalar inhomogeneous linear stochastic differential equations. As an application, some  $\mathbb{R}^n$  vector-valued inhomogeneous nonlinear stochastic differential equations are reduced to random differential equations, facilitating pathwise study of the solutions.

## 1 A Review of Stochastic Exponential Formulas

We first review some existing results about solution formulas for linear stochastic differential equations (SDEs) or for their integral formulations. Let  $(\Omega, \mathcal{F}, (\mathcal{F}t), \mathbf{P})$  be a standard stochastic basis. For the following stochastic integral equation

$$X_t = 1 + \int_0^t X_{s-} dZ_s, \quad (1.1),$$

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where  $Z$  is a semimartingale with  $Z_0 = 0$ , Doléans-Dade (1970) proved that the unique solution of (1.1) is given by

$$X_t = \exp \left\{ Z_t - \frac{1}{2} \langle Z^c, Z^c \rangle_t \right\} \prod_{0 < s \leq t} (1 + \Delta Z_s) e^{-\Delta Z_s}. \quad (1.2)$$

In the literature  $X$  is called *Doléans exponential (or stochastic exponential)* of  $Z$ , and is denoted by  $\mathcal{E}(Z)$ . The formula (1.2) is called the *Doléans (or stochastic) exponential formula*.

In an unpublished paper, Yoeurp and Yor (1977) proved the following result for the solution formula of scalar SDEs (see also Revuz-Yor (1999) and Protter (2005) for the case where  $Z$  is a continuous semimartingale, and Melnikov-Shiryaev (1996)) for the general case).

**Theorem 1.1 (Yoeurp and Yor 1977)** *Let  $Z$  and  $H$  be semimartingales, and  $\Delta Z_s \neq -1$  for all  $s \in [0, \infty]$ . Then the unique solution of the inhomogeneous scalar linear SDE*

$$X_t = H_t + \int_0^t X_{s-} dZ_s, \quad t \geq 0, \quad (1.3)$$

*is given by*

$$X_t = \mathcal{E}(Z)_t \left\{ H_0 + \int_0^t \mathcal{E}(Z)_{s-}^{-1} dG_s \right\}, \quad (1.4)$$

*where*

$$G_t = H_t - \langle H^c, Z^c \rangle_t - \sum_{0 < s \leq t} \frac{\Delta H_s \Delta Z_s}{1 + \Delta Z_s}. \quad (1.5)$$

Jaschke (2003) extended equation (1.3) to the case where  $H$  is an adapted cadlag process, not necessarily a semimartingale. He proved that in this case the solution of (1.3) is given by:

$$X_t = H_t - \mathcal{E}(Z)_t \int_0^t H_{s-} d(\mathcal{E}(Z)_s^{-1}). \quad (1.6)$$

By using (1.6) Jaschke (2003) obtained a new proof of (1.4).

On the other hand, Emery (1978) considered the following  $n \times n$  matrix-valued stochastic equation

$$U(t) = I + \int_0^t (dL(s))U(s-), \quad (1.7),$$

where  $I$  is an  $n \times n$  identity,  $L$  is a given  $n \times n$  matrix-valued semimartingale with  $L(0) = 0$ , such that  $I + \Delta L(s)$  is invertible for all  $s \in [0, \infty]$ , where  $\Delta L(s) = L(s) - L(s-)$ . Emery proved that the equation (1.7) admits a unique solution  $U$ , which is  $n \times n$  matrix-valued semimartingale. We call it the stochastic exponential of  $L$  and denote it by  $\mathcal{E}(L)$ . However, there is no explicit expression for such a stochastic exponential in general.

Jacod (1982) has studied the following inhomogeneous matrix-valued stochastic integral equation

$$X(t) = H(t) + \int_0^t (dL(s))X(s-), \quad (1.8)$$

where  $L$  is a given  $n \times n$  matrix-valued semimartingale with  $L0 = 0$ , such that  $I + \Delta L(s)$  is invertible for all  $s \in [0, \infty]$ , and  $H$  is an  $n \times m$  matrix valued semimartingale. For an  $n \times n$  matrix valued semimartingale  $A$  and an  $n \times m$  matrix valued semimartingale  $B$ , we let

$$[A, B](t) = \langle A^c, B^c \rangle(t) + \sum_{0 < s \leq t} \Delta A(s) \Delta B(s).$$

Here  $A^c$  denotes its continuous martingale part defined componentwise by  $(L^c)_j^i = (L_j^i)^c$ , and

$$\langle A^c, B^c \rangle_j^i = \sum_k \langle (A^c)^i_k, (B^c)_j^k \rangle. \quad (1.9)$$

Using these notations the result of Jacod (1982) implies the following

**Theorem 1.2 (Jacod 1982)** *The unique solution of (1.8) is given by*

$$X(t) = \mathcal{E}(L)(t) \left\{ H(0) + \int_0^t \mathcal{E}(L)(s-)^{-1} dG(s) \right\}, \quad (1.10)$$

where

$$G(t) = H(t) - \langle L^c, H^c \rangle(t) - \sum_{0 < s \leq t} (1 + \Delta L(s))^{-1} \Delta L(s) \Delta H(s). \quad (1.11)$$

In particular, if  $L$  and  $H$  are continuous semimartingales, an expression for the solution of (1.8) is given in Revuz-Yor (1999) as follows:

$$X(t) = \mathcal{E}(L)(t) \left\{ H(0) + \int_0^t \mathcal{E}(L)(s)^{-1} (dH(s) - d[L, H](s)) \right\}. \quad (1.12)$$

The objective of the present note is to generalize equation (1.8) to the case where  $H(t)$  is a given  $n \times m$  matrix-valued adapted cadlag process, not necessarily a semimartingale. We give an expression of the solution of (1.8) for this case and also give a simpler proof for Theorem 1.1. Our result extends (1.6) of Jaschke (2003) to matrix-valued case. As an application, we reduce some  $\mathbb{R}^n$ -valued inhomogeneous *nonlinear* SDEs to random differential equations (RDEs) — differential equations with random coefficients. This facilitates pathwise study of solutions and is an important step in the context of random dynamical systems approaches; see Arnold (1998).

## 2 Matrix-valued Inhomogeneous Linear SDEs

Léandre (1985) obtained the following result about stochastic equation (1.7). If we denote by  $V$  the inverse of  $\mathcal{E}(L)$ , then  $V$  is the solution of the following equation:

$$V(t) = I + \int_0^t V(s-)dW(s),$$

where

$$W(t) = -L(t) + \langle L^c, L^c \rangle(t) + \sum_{0 < s \leq t} (1 + \Delta L(s))^{-1} (\Delta L(s))^2.$$

That means  $\mathcal{E}(L)^{-1} = \mathcal{E}(W^\tau)^\tau$ . We refer the reader to Karandikar (1991) for a detailed proof of this result.

Now we will use this result of Léandre (1985) to solve the following inhomogeneous stochastic integral equation

$$X(t) = H(t) + \int_0^t (dL(s))X(s-), \quad (2.1),$$

where  $L$  is a given  $n \times n$  matrix-valued semimartingale with  $L_0 = 0$ , such that  $I + \Delta L(s)$  is invertible for all  $s \in [0, \infty]$ ,  $H(t)$  is a given  $n \times m$  matrix-valued adapted cadlag process.

Our main result is the following.

**Theorem 2.1** *The unique solution of (2.1) is given by*

$$X(t) = H(t) - \mathcal{E}(L)(t) \int_0^t (d\mathcal{E}(L)(s)^{-1})H(s-). \quad (2.2)$$

*If  $H$  is  $n \times m$  matrix-valued semimartingale, then  $X(t)$  has the same expression as given by (1.10) and (1.11).*

**Proof.** We denote  $\mathcal{E}(L)$  and  $\mathcal{E}(L)^{-1}$  by  $U$  and  $V$ , respectively. We are going to show that the process

$$X(t) = H(t) - U(t) \int_0^t (dV(s))H(s-),$$

defined by (2.2), satisfies equation (2.1). Since  $UV = I$ , by the integration by parts formula (see Karandikar (1991)) we get

$$\begin{aligned} 0 &= d(U(t)V(t)) = U(t-)dV(t) + (dU(t))V(t-) + d[U, V](t) \\ &= d(U(t)V(t)) = U(t-)dV(t) + dL(t) + d[U, V](t). \end{aligned}$$

Once again by the integration by parts formula, using the above result and the fact that  $dU(t) = (dL(t))U(t-)$ , we have

$$\begin{aligned} d(X(t) - H(t)) &= -U(t-)dV(t)H(t-) - (dU(t))\left(\int_0^{t-} dV(s)H(s-)\right) - (d[U, V](t))H(t-) \\ &= (dL(t))[H(t-) - U(t-)\left(\int_0^{t-} dV(s)H(s-)\right)] = (dL(t))X(t-). \end{aligned}$$

This shows that the process  $(X(t))$  defined by (2.2) satisfies (2.1).

Now we assume that  $(H(t))$  is an  $n \times m$  matrix-valued semimartingale. Using the notations in Section 1 we can verify that

$$G(t) = H(t) + [W, H](t). \quad (2.3)$$

By the integration by parts formula, using (2.3) and the fact that  $dV(t) = V(t-)dW(t)$ , we obtain

$$\begin{aligned} 0 &= d(V(t)H(t)) = V(t-)dH(t) + (dV(t))H(t-) + V(t-)d[W, H](t) \\ &= V(t-)dG(t) + dL(t) + (dV(t))H(t-), \end{aligned}$$

from which we get

$$H(t) = U(t) \left\{ H(0) + \int_0^t V(s-)dG(s) + \int_0^t dV(s)H(s-) \right\}.$$

Thus, if we let  $(G(t))$  be defined by (2.3) then  $(X(t))$  has the expression of (2.2), and consequently it satisfies equation (2.1). The proof of the theorem is complete.  $\square$

### 3 An Application to Nonlinear SDEs

In this section we will apply our results in Theorem 2.1 to reduce an inhomogeneous *nonlinear* SDE to a RDE (random differential equation). Now we consider the following  $n$ -dimensional nonlinear SDE (but with a linear multiplicative noise term):

$$dX^i(t) = f^i(t, X(t))dt + \sum_{j=1}^n C_j^i(t)X^j(t)dB^j(t), \quad X^i(0) = x^i, \quad (3.1)$$

where  $C(t)$  is an  $n \times n$  matrix-valued measurable function,  $f(t, x)$  is a  $\mathbb{R}^n$ -valued measurable function on  $[0, \infty) \times \mathbb{R}^n$ , and  $B(t) = (B^1(t), \dots, B^n(t))^\tau$  is a  $n$ -dimensional Brownian motion. Put

$$L_j^i(t) = \int_0^t C_j^i(s)dB^j(s), \quad i, j = 1, \dots, n; \quad H(t) = X(0) + \int_0^t f(s, X(s))ds. \quad (3.2)$$

Then (3.1) can be rewritten in the form of (2.1). For such  $L$ , the equation (1.9) is reduced to the following linear equation:

$$dU_j^i(t) = \sum_{k=1}^n C_k^i(t)U_j^k(t)dB^k(t), \quad U_j^i(0) = \delta_j^i, \quad (3.3)$$

In the present case we have  $G(t) = H(t)$ . So according to (2.2), the solution of (3.1) can be expressed as

$$X(t) = U(t) \left\{ x + \int_0^t U(t)^{-1} f(s, X(s))ds \right\}, \quad (3.4)$$

where  $U$  is the solution of (3.3). Unfortunately, even in this case we are not able to give an explicit expression for  $U(t)$ . Let  $Y(t) = U(t)^{-1}X(t)$ , then

$$Y(t) = \left\{ x + \int_0^t U(s)^{-1} f(s, U(s)Y(s)) ds \right\}. \quad (3.5)$$

This is the integral formulation of a RDE (random differential equation). Once we have sample path solution  $Y(t)$  of this transformed RDE, we obtain the solution of the original SDE via  $X(t) = U(t)Y(t)$ .

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